

Polynomial Simulations of Decohered Quantum Computers

D. Aharonov*

M. Ben-Or†

Abstract

Recently it has become clear, that a key issue in quantum computation is understanding how interaction with the environment, or “decoherence”, effects the computational power of quantum computers. We adopt the standard physical method of describing systems which are interwound with their environment by “density matrices”, and within this framework define a model of decoherence in quantum computation.

Our results show that the computational power of decohered quantum computers depends strongly on the amount of parallelism in the computation. We first present a simulation of decohered sequential quantum computers, on a classical probabilistic Turing machine, and prove that the expected slowdown of this simulation is polynomial in time and space of the quantum computation, for any non zero decoherence rate. Similar results hold for Quantum computers that are allowed to operate on logarithmic number of qubits at a time.

For decohered quantum circuits (with local gates), the situation is more subtle and depends on the decoherence rate, η . We find that our simulation is efficient for circuits with decoherence rate η higher than some constant η_1 , but exponential for a general (random) circuit subjected to decoherence rate lower than some constant η_2 . The transition from exponential cost to polynomial cost happens in a short range of decoherence rates. We use computer experiments to exhibit the phase transitions in various quantum circuits.

1 Introduction

Quantum Turing Machines[8, 4] and Quantum Circuits[7, 19] challenge the so called “polynomial Church thesis” which asserts that “randomized Turing machines can simulate with polynomial slowdown any computational device.” In particular Shor’s quantum factoring algorithm[16] provides within this theoretical framework, an efficient solution, to a problem for which no classical polynomial algorithm is known.

It is yet unclear whether and how quantum computers will be physically realizable,[12, 9] but as any physical system, they *in principle* will not be ideally isolated, and will to some extent interact with their environment. Such interaction causes the state of the computer to be interwound, or entangled, with the state of the environment, a process called “decoherence” [18, 17, 14]. A real challenge to the “polynomial church thesis” would come from *decohered* quantum computers. Early works[5] showed that the effects of *decoherence* on the quantum computation can not be ignored, and that decoherence may limit the applicability of quantum algorithms. Thus, understanding the computational power of decohered quantum computers is essential.

Decoherence takes us out of the standard model of quantum computers described by *pure states*, as the state of a decohered computer is in general a *mixed state*. Mixed states are used when there is only partial knowledge about the system, due to the fact that the system is interwound with its environment. Such a state can be represented as a probability p_k for the system to be in the pure state $|\alpha_k\rangle$. This description is not unique and instead we use a unique representation of mixed states called *density matrices*[11], to describe the state of the quantum computer[2].

In order to incorporate decoherence into the model, we need to add a clock to the circuit since the process of errors is dynamic and depends on time. A circuit is associated a *timing* which indicates the time step for each gate to operate. To add decoherence, the following assumptions are made: (1) Each qubit decoheres independently

*Institutes of Physics and Computer science, The Hebrew University, Jerusalem, Israel, E-mail: doria@cs.huji.ac.il

†Institute of Computer science, The Hebrew University, Jerusalem, Israel, E-mail: benor@cs.huji.ac.il

(single qubit faults), and (2) No decoherence takes place inside the gates. Each qubit decoheres, or undergoes a fault with probability η per step, and η is referred to as the *decoherence rate*. The triplet of a quantum circuit, a timing of the circuit, and a decoherence rate construct together what we call a *quantum medium*, which operates subjected to different types of single qubit faults. The list of times and places where faults had occurred, namely the *fault path*, is random, and naturally, the function that the medium computes is the weighted average over the outputs for each possible *fault path*. This model for decoherence is richer than it seems at first glance: it is equivalent to noise caused by weak interactions with the environment, and, to a model in which there are many possible kinds of faults that can occur. Our results therefore apply to these kinds of decoherences too.

In this paper we exhibit upper bounds on the computational power of quantum mediums subjected to faults. As a first observation, note that regarding upper bounds, it is enough to restrict the discussion to faults of a specific type, which we choose to be *collapses* of a single qubit, or in other words: measurements of a single qubit in some basis. It suffices to deal with collapses because of two reasons: The first is that if a quantum medium claims to compute a function fault tolerantly, it is supposed to compute it when subjected to any fault. The second, and maybe more significant reason, is that it is reasonable to assume that every physical realization of quantum computers will be subjected to some extent to collapses[17].

We present a simulation of a quantum medium subjected to single qubit collapses, on a classical probabilistic Turing machine. The simulation chooses, with the appropriate probability, a fault path. Then it keeps track of the development in time of the density matrix of the medium according to this path, and gets a final density matrix. The simulation imitates measurements on this density matrix and outputs the result. Since the algorithm chooses the path with the correct weight, the distribution over the results will be as that of the medium. How does the simulation keep track on the development of the density matrix, which is exponential in size? Trying to be efficient, the simulation presents this matrix as a tensor product of smaller density matrices. The input string is described as a tensor product of the density matrices of each qubit, but if no faults occurred, the computer can soon entangle all the qubits together, with no way to view their exponential density matrix as a tensor product of smaller matrices. Collapses prevent this from happening, since after a qubit collapses to one of its basic states it is no longer entangled to the other qubits, and can be described by its own density matrix.

We are interested in the expected cost of the simulation. The results emphasize the significance of the amount of parallelism in the computation. To gain some intuition, one can view the decohered quantum computation as a struggle between two forces: The computation, which tries to use the exponential dimension of the Hilbert space, by entangling as many qubits as it can, and the decoherence, which destroys this entanglement by collapses. From this point of view, a sequential computer obviously can not win the battle - each time step it applies only one local operation, but $\theta(n)$ qubits collapse. The picture in mediums with general parallelism is different, since both forces that compete have linear power: $\theta(n)$ gates can be applied each time step, and the decoherence collapses a percentage of the n qubits. Since the competition here is even, the question which force wins in this struggle is delicate and requires careful consideration.

Our results show that:

- The expected slowdown of our simulation applied on **mediums with $O(\log(n))$ parallelism**, with any non zero decoherence rate, is polynomial in the number of qubits and time steps. This holds also for multi head Quantum Turing machines that are allowed to operate on $O(\log(n))$ qubits each time step.

The mathematical techniques involved in the case of **mediums with general parallelism** are more sophisticated. In our analysis we assume that the gates are of fan-in not bigger than two[6], but the results can be extended to any constant fan-in:

- **High decoherence rate:** The simulation of mediums with decoherence rate higher than some constant is efficient. Single-qubit-gates can not entangle different qubits, and therefore do not participate in the “struggle” against decoherence. Hence if two-qubits-gates are applied “rarely”, the cost is polynomial already at lower decoherence rates.

- **Low decoherence rates:** The cost of simulating a random circuit is exponential, for decoherence rates lower than some constant.

- **Phase transition in the cost:** Results regarding phase transitions[10] imply that the cost of simulating a random circuit transforms from exponential to polynomial in a very short range of decoherence rates. As for other quantum circuits, if the cost transforms from exponential to polynomial this transition is sharp (by extension of [10], private communication).

We used computer experiments to show that in the case of a random circuit, the phase transition occurs at

$\eta_0 \approx 0.63$. We also looked at a medium which is a one dimensional array of qubits, where gates are applied only on nearest neighbors, and found that in this case the transition occurs already at $\eta \approx 0.50$.

Organization of paper: In section 2 we recall the basic definitions of quantum circuits and mixed states. In section 3 we define the model of quantum mediums with faults. Section 4 is devoted to the simple case of simulating sequential mediums. In section 5 we concentrate on simulating mediums with general parallelism. Section 6 discusses consequences of this work.

2 Quantum circuits with mixed states

In this section we recall some physical definitions of quantum systems of n two-state particles, Hilbert space, computational basis, pure states, mixed states and density matrices[11]. We then describe the model of quantum circuits with mixed states, defined in [2].

2.1 Pure states and mixed states

Pure states: We deal with systems of n two-state quantum particles, or “qubits”. The *pure state* of such a system is a unit vector, denoted $|\alpha\rangle$, in the Hilbert space¹ \mathcal{C}^{2^n} i.e. a 2^n dimensional complex space. We view \mathcal{C}^{2^n} as a tensor product n two dimensional spaces, each corresponding to a qubit: $\mathcal{C}^{2^n} = \mathcal{C}^2 \otimes \dots \otimes \mathcal{C}^2$. As a basis for \mathcal{C}^{2^n} , we use the 2^n orthogonal *basic states*: $|i\rangle = |i_1\rangle \otimes |i_2\rangle \dots \otimes |i_n\rangle$, $0 \leq i < 2^n$, where i is in binary representation. and each i_j gets 0 or 1. Such a state corresponds to the j 'th qubit being in the state $|i_j\rangle$. A pure state $|\alpha\rangle \in \mathcal{C}^{2^n}$ is a *superposition* of the basic states: $|\alpha\rangle = \sum_{i=1}^{2^n} c_i |i\rangle$, with $\sum_{i=1}^{2^n} |c_i|^2 = 1$. $|\alpha\rangle$ corresponds to the vector $v_\alpha = (c_1, c_2, \dots, c_{2^n})$. v_α^\dagger , the complex conjugate of v_α , is denoted $\langle\alpha|$. The inner product between $|\alpha\rangle$ and $|\beta\rangle$ is $\langle\alpha|\beta\rangle = (v_\alpha, v_\beta)$. The matrix $v_\alpha^\dagger v_\beta$ is denoted as $|\alpha\rangle\langle\beta|$. An isolated system of n qubits develops in time by a unitary matrix² of size $2^n \times 2^n$: $|\alpha(t_2)\rangle = U|\alpha(t_1)\rangle$. A quantum system in \mathcal{C}^{2^n} can be *observed* by *measuring* the system. An important measurement is a *basic measurement* of a qubit q , of which the possible outcomes are 0, 1. For the state $|\alpha\rangle = \sum_{i=1}^{2^n} c_i |i\rangle$, the probability for outcome 0 is $p_0 = \sum_{i, i_q=0} |c_i|^2$ and the state of the system will *collapse* to $|\beta\rangle = \frac{1}{p_0} \sum_{i, i_q=0} c_i |i\rangle$, (the same for 1). In general, an *observable* O over \mathcal{C}^{2^n} is an hermitian³ matrix, of size $2^n \times 2^n$. To apply a measurement of O on a pure state $|\alpha\rangle \in \mathcal{C}^{2^n}$. write $|\alpha\rangle$ uniquely as a superposition of unit eigenvectors of O : $|\alpha\rangle = \sum_i c_i |o_i\rangle$, where $|o_i\rangle$ have different eigenvalues. With probability $|c_i|^2$ the measurement's outcome will be the eigenvalue of $|o_i\rangle$, and the state will *collapse* to $|o_i\rangle$. A unitary operation U on k qubits can be applied on n qubits, $n \geq k$, by taking the extension \tilde{U} of U , i.e. the tensor product of U with an identity matrix on the other qubits. The same applies for an observable O to give \tilde{O} .

Mixed states: A system which is not ideally isolated from it's environment is described by a *mixed state*. There are two equivalent description of mixed states: mixtures and density matrices. Two different density matrix can be distinguished by a measurement, and two systems with the same density matrix are quantumly indistinguishable. In contrast, the mixture description is not unique: different mixtures may yield the same density matrix. We use density matrices in this paper.

mixtures: A system in the **mixture** $\{\alpha\} = \{p_k, |\alpha_k\rangle\}$ is with probability p_k in the pure state $|\alpha_k\rangle$. The rules of development in time and measurements for mixtures are obtained by applying **classical** probability to the rules for pure states. A unitary matrix U transforms a mixture $\{p_k, |\alpha_k\rangle\}$ to $\{p_k, U|\alpha_k\rangle\}$. To apply an observable O on the mixture $\{\alpha\} = \{p_j, |\alpha_j\rangle\}$. write each pure state as a sum of eigenvectors of O with *different* eigenvalues: $|\alpha_j\rangle = \sum_i c_j^i |o_j^i\rangle$. The probability to get an eigenvalue λ_i is $Pr(\lambda_i) = \sum_j p_j |c_j^i|^2$. In the resulting mixture, under the condition the result was λ_i , each pure state has collapsed to it's part that is consistent with λ_i , and the conditioned probabilities are computed classically: $\{\alpha\}_{\lambda_i} = \{\frac{|c_j^i|^2 p_j}{\sum_j |c_j^i|^2 p_j}, |o_j^i\rangle\}$. One can also apply an unconditioned measurement, i.e. not condition on the outcome of the measurement but instead average on the possible outcomes. This gives the mixed state $O \circ \{\alpha\} = \{Pr(\lambda_i), \{\alpha\}_{\lambda_i}\}$.

Density matrices: A density matrix ρ on \mathcal{C}^{2^n} is an hermitian positive semi definite complex matrix of dimensions $2^n \times 2^n$, with $tr(\rho) = 1$. A pure state $|\alpha\rangle = \sum_i c_i |i\rangle$ is associated the density matrix $\rho_{|\alpha\rangle} = |\alpha\rangle\langle\alpha|$ i.e. $\rho_{|\alpha\rangle}(i, j) =$

¹ A Hilbert space is a vector space with an inner product

² Unitary matrices preserve the norm of any vector and satisfy the condition $U^{-1} = U^\dagger$

³ An hermitian matrix H satisfies $H = H^\dagger$

$c_i c_j^*$. A mixture $\{\alpha\} = \{p_l, |\alpha_l\rangle\}$, is associated the density matrix : $\rho_{\{\alpha\}} = \sum_l p_l \rho_{|\alpha_l\rangle}$. The operations on a density matrix are defined such that the correspondence to mixtures is preserved. If a unitary matrix U transforms the mixture $\{\alpha\} = \{p_l, |\alpha_l\rangle\}$ to $\{\beta\} = \{p_l, U|\alpha_l\rangle\}$. Then $\rho_{\{\beta\}} = \sum_l p_l U|\alpha_l\rangle\langle\alpha_l|U^\dagger = U\rho_{\{\alpha\}}U^\dagger$. Applying a measurement of an observable O on ρ , which is written in a basis of eigenvectors v_i of O . gives, the probability for an outcome λ is the sum of the diagonal terms of ρ , which relate to the eigenvalue λ : $pr(\lambda) = \sum_{i=1}^{2^n} \rho_{v_i, v_i} \delta(\lambda_i = \lambda)$. conditioned that the outcome is the eigenvalue λ , the resulting density matrix is $O_\lambda \circ (\rho)$, which we get by first putting to zero all rows and columns in ρ , which relate to eigenvalues different from λ , and then renormalizing this matrix to trace one. Without conditioning on the outcome the resulting density matrix will be $O \circ (\rho) = \sum_k Pr(\lambda_k) O_{\lambda_k} \circ (\rho)$. which differs from ρ , only in that the entries in ρ which connected between different eigenvalues are put to zero. Given a density matrix ρ of n qubits, the reduced density matrix of a subsystem, A , of, say, m qubits is defined as an average over the states of the other qubits: $\rho|_A(i, j) = \sum_{k=1}^{2^{n-m}} \rho(ik, jk)$.

2.2 Quantum circuits with mixed states

We describe the model of quantum circuits[7, 19], with mixed states[2]: A *quantum unitary gate* of order k is a complex unitary matrix of size $2^k \times 2^k$. A density matrix ρ will transform by the gate to $g \circ \rho = \tilde{U}\rho\tilde{U}^\dagger$, where \tilde{U} is the extension of U . Using density matrices one can also define a non unitary gate: A *measurement gate* of order k is a complex hermitian matrix of size $2^k \times 2^k$. A density matrix ρ will transform by the gate to $g \circ \rho = \tilde{O} \circ (\rho)$. A *Quantum circuit* is a directed acyclic graph with n inputs and n outputs. Each node v in the graph is labeled by a quantum gate g_v . The in-degree and out-degree of v are equal to the order of g_v . Some of the outputs are labeled “result” to indicate that these are the qubits that will give the output of the circuit. The wires in the circuit correspond to qubits. An initial density matrix ρ transforms by a circuit Q to a final density matrix $Q \circ \rho = g_t \circ \dots \circ g_2 \circ g_1 \circ \rho$, where the gates $g_t \dots g_1$ are applied in a topological order. For an input string i , the initial density matrix is $\rho_{|i\rangle}$. The output of the circuit is the outcome of applying basic measurements of the result qubits, on the final density matrix $Q \circ \rho_{|i\rangle}$. Since the outcomes of measurements are random, the function that the circuit computes is a *probabilistic function*, i.e. for input i it outputs strings according to a distribution which depends on i .

3 Quantum mediums with faults

In this section we define quantum mediums subjected to probabilistic single qubit faults of one type. We then show that the probabilistic fault model is equivalent to a deterministic weak fault model and to a model with many possible types of faults.

3.1 Quantum mediums with probabilistic faults

In order to incorporate faults into quantum circuits, we need to add a clock, since the process of errors depends on time. A quantum circuit will be associated a *timing*, which indicates the time step each gate is applied. The timing should be consistent with the circuit: if a gate gets as input output of another gate, then it should be applied at least one time step after the other gate. In a faultless computation, one can assume without loss of generality that all the qubits are initialized and output together. In the presence of faults, it is still not clear whether allowing to input constants at any time helps the computation, since using qubits for a limited time might protect them against faults. Taking the stronger model, we permit also that qubits are input and output at different times, and we say a qubit is *alive* from t_1 to t_2 if it is input to the circuit at t_1 and output at t_2 . We will assume all timings start at 0, and denote by T the last time step.

We would like to incorporate in the model single qubit faults. With out loss of generality, a single qubit fault, F , operating on the q' th qubit in the density matrix ρ of n qubits, can be described by adding a qubit to the system, that represents the environment, and letting a quantum gate to operate on it and the q' th qubit, and then taking the reduced density matrix to the original qubits. The fault is thus specified by the gate. The resulting density matrix is denoted as $F^q \circ \rho$. In this model, a live qubit undergoes the fault F with some probability η per time step, and η is called the *decoherence rate*. (One can allow the fault to depend on the time and on the exact qubit. All the results of this paper still hold.)

We now define the medium that realizes a quantum circuit with a certain timing, and is subjected to a certain decoherence rate:

definition 1 A quantum medium $M(Q, \tau, \eta)$ is a triplet: A quantum circuit Q , a timing τ on Q , and a decoherence rate $0 \leq \eta \leq 1$.

We define a fault path, which is a possible list of where and when the faults occur:

definition 2 Let $M(Q, \tau, \eta)$ be a medium of n qubits, which lives T steps. Denote by t_1^q, t_2^q the times of berth and death of the q 'th qubit. A fault path is a set of different pairs (q_j, t_j) , where $1 \leq q_j \leq n$, $t_1^{q_j} \leq t_j \leq t_2^{q_j}$.

We associate with each path a weight, which is the probability for it to occur:

definition 3 Let $M(Q, \tau, \eta)$ be a medium of n qubits which has V pairs (q, t) where faults can occur. The weight of a path σ is $w(\sigma) = d^{|\sigma|}(1 - d)^{V - |\sigma|}$.

M computes a probabilistic function, where the distribution $f(i)$ which M computes for an input i is defined as a weighted average over the outputs which the final density matrix of each fault path generates. To formalize this, we need notions of development in time of the medium. For a medium of n qubits, an evolution is a possible such development, i.e. a list $\{E(i)\}_{i=0}^T$ of density matrices of n qubits. There are special evolutions which develop subjected to a certain fault path, and besides that “obey” the gates:

definition 4 let M be a medium of n qubits, F a fault, σ a fault path. An evolution E is called a “ σ, F -trajectory of M ” if it satisfies $E(t + 1) = F^{\sigma_t} \circ \prod_k g_t^k \circ E(t)$ where g_t^k are the gates applied at time t , and F^{σ_t} are faults applied in σ at time t .

We can now define the function computed by the medium when subjected to faults F :

definition 5 Let the medium M be subjected to faults F . For an input string i let $\rho_i = \sum_{\sigma} w(\sigma) E_i^{\sigma}(T)$, where E_i^{σ} is an F, σ -trajectory of M which satisfies $E_j^{\sigma}(0) = \rho_{|i\rangle}$. The probability for M to output a string j for the input i is the probability to get an outcome j when applying basic measurements on the result qubits in ρ_i .

3.2 Equivalence to a weak interactions

In the weak interactions model, each qubit undergoes a *weak fault*, deterministically, each time step. The weak fault operates on a density matrix in the following way: $WF^q \circ \rho = (1 - \eta)\rho + \eta F^q \circ \rho$. where F is a single qubit fault, and η is the strength of the interaction with the environment. The first term serves as a guard, so that the change in the density matrix is limited. To show that this model is equivalent to the probabilistic model, it suffices to show that there is no difference between a mixed state which is a weighted sum of mixed states, and a mixed state prepared by choosing, with the correct probability, one of the mixed states in the sum.

lemma 1 Let $\rho = p_1\rho_1 + p_2\rho_2$. The mixed state ρ is quantumly equivalent to a mixed state prepared by choosing with probability p_i the density matrix ρ_i .

Proof: The distinguish ability of quantum states is by measurements. By definition, the statistics of any measurement will be the same in the two cases. ■

3.3 Equivalence to a model with many possible faults

We can also define a model which seems more physical: At each time step, and for each live qubit, different possible faults F_i , can occur with probabilities η_i , where $\eta = \sum_i \eta_i$. It might seem surprising at first sight, but this model is equivalent to our probabilistic model since we can define the one fault in our model to be a probabilistic combination of all the faults: $F \circ \rho = \frac{1}{\eta} \sum_i \eta_i F_i \circ \rho$.

4 Polynomial simulations of sequential mediums

We describe a simulation of a quantum computation by a quantum medium, on a randomized classical Turing machine. As described in the introduction, it suffices for our purposes to regard only collapse faults. We show that the expected slowdown of the simulation, for sequential mediums and mediums which operate gates only on $O(\log(n))$ qubits each time step, is *polynomial* in the time and number of qubits of the medium, for any non-zero decoherence rate, $\eta > 0$.

To study the effects of decoherence on QTM one must refer to some possible physical realizations of QTMs. Our results apply immediately to the possible realization of QTM by sequential Quantum circuits, suggested by Yao[19]. At the end of this section we informally describe how to extend the results to general multi-head QTM.

4.1 A classical simulation of quantum mediums

We define an algorithm $A(M, F)(i)$ which gets as an input a description of a Quantum medium M , a single qubit observable F , and an input string i , and outputs the function that the medium outputs when subjected to collapses of type F . A presents a density matrix, ρ , of n qubits, by a *configuration*: a list of density matrices, ρ_j , each ρ_j describing a set of entangled qubits, or a “cluster”, such that there is no entanglement between different clusters. ρ is the tensor product of the matrices in the configuration. let us describe the probabilistic algorithm A :

Input:

- Description of a medium $M = (Q, \tau, d)$,
- string of n qubits, i .
- The eigenstates $|e_1\rangle, |e_2\rangle$, of the observable F .

Algorithm:

- A chooses a decoherence path σ with probability $w(\sigma)$, according to the decoherence rate η .
- A initializes the configurations $conf(0), conf^*(0)$ to be a list of matrices of one qubit, $|0\rangle\langle 0|$ or $|1\rangle\langle 1|$ that correspond to i .
- For $t=1$ to T , A does:
 - a) To simulate the computation step: Let g be a gate that operates at time t . Let ρ^{q_i} be the matrices in $conf(t)$ that describe qubits on which g operates. A replaces them in the configuration by the matrix $g \circ (\rho^{q_1} \otimes \dots \rho^{q_n})$. For two qubits in the same cluster described by ρ^q , we take ρ^q in the tensor product only once. This is done for all the gates of the t' th time step to give $conf^*(t)$.
 - b) To simulate a decoherence step: Let (q, t) be a pair in σ . Let ρ_q be the matrix in $conf^*(t)$ that describes the qubit q , and maybe a set B of more qubits. A computes, from ρ_q , the probability for q to collapse on each of the eigenvalues of F . Then A tosses a coin with this bias, to decide on which of the eigenstates the qubit collapses. Let the result be $|e_1\rangle$. A replaces ρ_q in $conf^*(t)$ by the matrix $|e_1\rangle\langle e_1|$, and the reduced matrix $F_{e_1} \circ (\rho_q)|_B$. This simulates a measurement of the q' th qubit, *conditioned* that the outcome was e_1 .

Output: To output the results of the computation, A imitates basic measurements of the result qubits, as in the decoherence step, and outputs the outcome.

To show that A computes the same function as the quantum medium, we observe that the simulation actually computes the F, σ – *trajectory* of the medium, for the path it chose, except one thing: The simulation makes a *conditioned* measurement, and continues according to the result, while the effect of the measurement faults is of an *unconditioned* measurement. However:

lemma 2 *The output distribution of $A(M, F)(i)$ equals the distribution $f_M(i)$ that M computes when subjected to faults F , for an input i .*

Proof: By lemma 1, the mixed state $F^q \circ \rho = P_1|e_1\rangle\langle e_1| \otimes F_{e_1} \circ \rho|_B + P_2|e_2\rangle\langle e_2| \otimes F_{e_2} \circ \rho|_B$ can not be distinguished from a mixed state prepared from $|e_i\rangle\langle e_i| \otimes F_{e_i} \circ \rho|_B$ with probability p_i . Therefore the effect of the unconditioned measurement is the same as taking with the correct probabilities the conditioned measurements. ■

4.2 The simulation is efficient.

To analyze the cost of the simulation, we divide the qubits each time step to two sets: The *individual qubits*, that in the configuration of A at time t , are described by their own density matrix, and the other qubits, which are entangled between themselves. Each individual qubit is described in the configuration by a density matrix with 4 entries, but the description of the entangled qubits might be exponential in their number. Denote by $K(t)$, $K^*(t)$ the sizes of sets of the non-individual qubits at time t in $conf(t)$, $conf^*(t)$ in the simulation. To bound the cost of the simulation from above, we first prove that the probability for the number of non-individual qubits to be large is exponentially small.

lemma 3 *Let M be a medium which operates on no more than P qubits each step, subjected to collapses of type F , with $\eta > 0$. Consider $A(M, F)(i)$. There exist constants $a > 4$, and $b > 0$ such that $Pr(K(t) \geq bP + j) \leq \frac{1}{a^j}$, $Pr(K^*(t) \geq bP + j + P) \leq \frac{1}{a^j}$.*

Proof: We use induction on t . For $t = 0$, $K(0) = 0$, $K^*(0) \leq P$. For the induction step, assume $Pr(K(t) \geq bP + j) \leq \frac{1}{a^j}$ and $Pr(K^*(t) \geq bP + j + P) \leq \frac{1}{a^j}$ for t . To analyze $Pr(K(t+1))$, We apply decoherence on $K^*(t)$. The probability to reduce the number of non-individual qubits is binomial so:

$$\begin{aligned} Pr(K(t+1) \geq bP + i) &= \sum_{j=0}^{\infty} Pr(K(t+1) = bP + i + j) = \\ &\leq \sum_{j,l=0}^{\infty} \frac{1}{a^{i+j+l-P}} \binom{bP + i + j + l}{l} \eta^l (1-\eta)^{bP+i+j} = \\ &= \sum_{j=0}^{\infty} \frac{(1-\eta)^{bP+i+j}}{a^{i+j-P}} \left(\sum_{l=0}^{\infty} \binom{bP + i + j + l}{l} \left(\frac{\eta}{a}\right)^l \right) = \\ &= \frac{a^{P+1}}{a-1} \left(\frac{(1-\eta)a}{a-\eta} \right)^{bP} \left(\frac{1-\eta}{a-\eta} \right)^i. \end{aligned}$$

We want this to be smaller than $\frac{1}{a^i}$, for any i . This implies two demands: (1) $\frac{1-\eta}{a-\eta} \leq \frac{1}{a}$, which is achieved for $a \geq 1$. (2) The coefficient of the geometrical sequence must be smaller than 1. There exists b that fixes that since $\frac{(1-\eta)a}{a-\eta}$ is strictly smaller than 1, because $\eta > 0$.

For $Pr(K^*(t+1))$ we have that $K^*(t+1) \leq K(t+1) + P$. Then $Pr(K^*(t+1) \geq bP + i + P) \leq Pr(K(t+1) \geq bP + i) \leq \frac{1}{a^i}$. ■

We can now prove that the expected number of entries of density matrices that A writes is polynomial:

lemma 4 *Let $M(Q, \tau, \eta)$ be a medium of n qubits with parallelism $O(\log(n))$, $\eta > 0$, F a measurement fault, i an input string. Consider $A(M, F)(i)$. The expected number of entries of the density matrices in all the configurations that A writes is polynomial in n and T .*

Proof: We use the definitions of the constants a and b from lemma 3. We compute the expectation value of $m(t)$, the number of entries for the matrices of the $K(t)$ non-individual qubits, at time t : $E(m(t)) = \sum_{m=1}^{\infty} Pr(m(t) = m)m \leq \sum_{j=0}^{\infty} Pr(K(t) = j)4^j \leq \sum_{j=0}^{bP-1} 4^j + \sum_{l=0}^{\infty} \frac{1}{a^l} 4^{bP+l}$, where the last inequality is by lemma 3. The sum is finite since $a > 4$. The expectation is constant if the parallelism, P , is constant, and for $P = c \log(n)$, the expectation is $\theta(n^{2cb})$. At time t we also have to write 4 entries for each small density matrix, which adds at most $4n$ entries. The expected total number of entries, (during the whole simulation) is the sum of the expectations in each time step. Thus it is polynomial in n and linear in T . ■

Theorem 1 *Let M be a medium of n qubits, life length T , with $\eta > 0$, and $O(\log(n))$ parallelism, F a measurement fault. The expected cost of $A(M, F)(i)$ will be polynomial in T and n .*

Proof: By lemmas 3,4 and the fact that one entry of a density matrix can be written in $\text{poly}(n, T)$ bits, if the specification of the gates is by $\text{poly}(n, T)$ bits. ■

4.3 Simulating quantum Turing machines

We briefly mention how to extend the above result to other possible models of QTM. We assume a decoherence model for multi-head QTM where besides single site faults, we allow faults on the heads' positions and the processor's state. Again we refer to the model of random collapses, where we can show:

Theorem 2 *Let Q be a QTM, subjected to measurement faults F of the sites, the heads' position and the machine's state, with $\eta > 0$. A classical probabilistic Turing machine can simulate T computation steps of Q on inputs of size n in time polynomial in T and n .*

Proof: Sketch. After every constant number of steps, the head's position will collapse to one place. Since at each time step it can only triple the number of sites it works on (by going Left, right or staying in place) the expected number of different paths of the head at each moment will remain constant. The simulation will remain efficient when the number of heads is $O(\log n)$ since then the number of paths is polynomial.

5 Simulating mediums with general parallelism

In this section we study the effects of decoherence on the computational power of general Quantum mediums, by applying the simulation A , defined in section 4.1, for mediums with general parallelism. We assume gates of fan-in ≤ 2 , but the results hold for any constant fan-in. We show analytically that the cost of the simulation has an expected polynomial slowdown for any medium with high enough decoherence. If the medium uses gates on more than one qubit not too frequently, the simulation is still polynomial even for lower decoherence rates. The simulation is proved to be exponential for a general random medium with low enough decoherence rate. Finally we present computer experiments, showing a phase transition in the expected cost at some decoherence rate, where the cost "jumps" from polynomial to exponential. These transition points are found for random circuits and nearest neighbor circuits. Known results [10] imply that for random circuits the transition happens in a range of decoherence rates which is $O(\frac{1}{\log(n)})$. For any sequence of quantum circuits for which there is a gap between the costs for high and low decoherences, the transition happens in a range which goes to zero as the size of the circuit goes to infinity.

5.1 Mediums with high decoherence rate.

It is proved that for high enough decoherence, the expected slowdown of the simulation A is polynomial. We need a more refined analysis than in the sequential case, since the line of reasoning applied there leads in the general parallelism case to a trivial exponential upper bound on the cost. Recall that a cluster in a configuration of A is a set of qubits described by the same density matrix in the configuration. In the analysis of the sequential case we referred to the total number of non-individual qubits, i.e qubits which are in clusters of size bigger than one. Here we take a closer look and consider the sizes of each cluster alone. We define $K_q(t), K_q^*(t)$ as the size of q 's cluster in $conf(t), conf^*(t)$, respectively. As long as the sizes of the clusters are $O(\log(n))$, the simulation is polynomial in n , since the number of entries in each density matrix in $conf(t)$ will be polynomial.

We intend to show by induction on t that the probability for $K_q(t)$ to be large decays exponentially. The difficulty here is to bound the change in the clusters' sizes during the computation step. If we could bound this, so that the clusters do not grow too much during the computation step, and if the decoherence is high enough to shrink them back down, then the induction step will work. In the following we analyze separately the change in the distribution of sizes of clusters, during the computation step, and during the decoherence step. Then we combine the two to show that the cost of the simulation is polynomial for high enough decoherence.

5.1.1 Clusters' Growth in a computation step

Assuming an exponential decaying bound on the sizes of clusters before the computation step, we can give an exponential decaying bound on the sizes of clusters after the computation step.

To do that we observe a connection to a branching process. Let the clusters in the configuration at time t , $conf(t)$, be A, B, C, \dots . Let M be the next computational step. The question is what is the cluster's size of a qubit

$a \in A$ after M is applied. Recall that we assume gates of fan-in two at the most. If a gate in M operates on two qubits then they are called *mates*. We associate with a a *tree* of which a is the *root*. a 's sons in the tree are the qubits in A . Suppose a_1 is a son of a and a mate of $b \in B$. Then all members in B are the sons of a_1 , and so on. In each stage we ignore qubits that already appear in the tree. The distribution on sizes of clusters corresponds to a distribution on the number of sons in a branching process. The distribution on the sizes of clusters after the computation corresponds to the sizes of trees in the branching process. If a has no mate, $K_a^*(t)$, which is the size of a 's cluster after M is applied, equals the size of a 's tree. If a has a mate, c , then $K_a^*(t)$ is the sum of sizes of a 's tree and of c 's tree, where again duplicated qubits are ignored.

We first consider a simplified version of the above process where (1) All gates in M are of fan-in 2, and (2) Gates in M operate only on qubits from different clusters. With these simplifications, the process that generates a tree, defined above, turns out to be a simple branching process, for which we can compute the probability distribution over the total size of two trees, given the distribution on the number of sons.

lemma 5 *Let B be a branching process, with $\Pr(z = j) \leq \frac{1}{a^j}, a > 8$ for z , the number of sons. Let T_a, T_b , be the sizes of two independent trees generated by B . Let $L = 2T_a + 2T_b - 2$. Then $\Pr(L = i) \leq a(\sqrt{2} - 1)^2(\sqrt{\frac{8}{a}})^i$.*

Proof: For the standard proofs in the following discussion, consult [13]. Let the generating functions⁴ of z, T, L be G_z, G_T, G_L . In the following formulas, the first[13] connects the generating functions of the size of a tree T to the number of sons Z . By definition of L we have. The second we have by definition of L :

$$G_{L+2}(x) = (G_T(x^2))^2, \quad G_T(x) = xG_z(G_T(x)) \quad (1)$$

We don't know the generating function for the variable z since we only have a bound on the distribution. Let us refer to the exponential decaying bound as a pseudo distribution, although it is not normalized to one, and define "pseudo generating functions", (denoted by a tilde), such that $\tilde{G}_z(x)$ will "generate" the exponential decay $\frac{1}{a^i}$: $\tilde{G}_z(x) = \sum_{i=0}^{\infty} \frac{1}{a^i} x^i = \frac{a}{a-x}$, and $\tilde{G}_T(x), \tilde{G}_L(x)$ will satisfy equations 1: $\tilde{G}_T(x) = x\tilde{G}_z(\tilde{G}_T(x))$, $\tilde{G}_{L+2}(x) = (\tilde{G}_T(x^2))^2$. The coefficients z_i, t_i, l_i in the analytic expansion of $G_z(x), G_T(x)$, and $G_{L+2}(x)$ are smaller than the corresponding coefficients $\tilde{z}_i, \tilde{t}_i, \tilde{l}_i$ of the "pseudo generating function," when the expansions exist. To prove this, note that the assumptions of the lemma imply $z_i \leq \tilde{z}_i$. Writing equation 1 using analytic expansions, one can show by induction that $t_i \leq \tilde{t}_i$. Using this and writing equation 1 in analytic expansion, we have $l_i \leq \tilde{l}_i$.

We intend to give an upper bound on \tilde{l}_i , since this will imply an upper bound on $l_{i+2} = \Pr(L = i)$ and therefore on $\Pr(L \geq i)$. Given $\tilde{G}_z(x)$, we can solve $\tilde{G}_T(x) = x\tilde{G}_z(\tilde{G}_T(x))$ for $\tilde{G}_T(x)$, which gives a quadratic equation. We choose the "minus" solution of the quadratic equation, which corresponds to the correct choice in the case of $G_T(x)$. It gives $\tilde{G}_T(\frac{a}{8}) = \frac{(\sqrt{2}-1)a}{\sqrt{8}}$, which implies

$$\frac{(\sqrt{2}-1)^2 a^2}{8} = (\tilde{G}_T(\frac{a}{8}))^2 = \tilde{G}_{L+2}(\sqrt{\frac{a}{8}}). \quad (2)$$

$\tilde{G}_T(x)$ is analytic for $|x| < \frac{a}{4}$, so $\tilde{G}_{L+2}(y)$ is analytic at $y = \sqrt{\frac{a}{8}}$, and we can write: $\frac{(\sqrt{2}-1)^2 a^2}{8} = \sum_{i=0}^{\infty} \tilde{l}_i (\sqrt{\frac{a}{8}})^i$, which gives $\tilde{l}_i \leq (\sqrt{2}-1)^2 \frac{a^2}{8} \left(\sqrt{\frac{8}{a}}\right)^i$ that implies the desired result. ■

We use the connection to the branching process defined above, to show an upper bound on the clusters' size after the computation step, given the upper bound before the computation step.

lemma 6 *Let $a > 8$, and $C = (\sqrt{2}-1)^2 \frac{a^{\frac{3}{2}}}{\sqrt{a}-\sqrt{8}}$. $\Pr(k_q(t) \geq j) \leq \frac{1}{a^{j-1}}$ implies $\Pr(K_q^*(t) \geq i) \leq C(\sqrt{\frac{8}{a}})^i$.*

Proof: Compare our process to the simplified branching process, where identifying $k_q(t) - 1$ with the variable z and $k_q^*(t)$ with L . Our process differs in two aspects: It might connect mates inside a cluster, and a qubit might not have a mate at all, i.e might not have a chance to give birth to children. The only effect of these fact can be to reduce the final cluster. ■

⁴The generating function $G_A(x)$ of a random positive integer A , is the expectation of x^A : $E(x^A) = \sum_{i=0}^{\infty} \Pr(A = i)x^i$

5.2 The decoherence step

We now show that with strong enough decoherence rate, the distribution over sizes of clusters is “pulled” by decoherence back below the original exponential decaying bound:

lemma 7 $\exists \eta_1 < 1, a > 8$ such that $\forall \eta > \eta_1 \Pr(K^*(t) \geq i) \leq C(\sqrt{\frac{8}{a}})^i$ with $C = (\sqrt{2} - 1)^2 \frac{a^{\frac{3}{2}}}{\sqrt{a} - \sqrt{8}}$, implies $\Pr(k(t+1) \geq l) \leq \frac{1}{a^{l-1}}$.

Proof: We find $\eta_1 < 0.97$. We have to calculate how a distribution over sizes of clusters changes because of the decoherence. The probability to reduce the number of qubits in a cluster is binomial so:

$$\begin{aligned}
\Pr(k(t+1) \geq l) &= \sum_{r=0}^{\infty} \Pr(k(t+1) = l+r) = \\
&\sum_{r=0}^{\infty} \sum_{p=0}^{\infty} \Pr(K^*(t) = l+r+p) \binom{l+r+p}{p} \eta^p (1-\eta)^{l+r} \leq \\
&\sum_{p=0}^{\infty} \Pr(K^*(t) \geq l+p) \sum_{r=0}^{\infty} \binom{l+r+p}{p} \eta^p (1-\eta)^{l+r} \leq \\
&\leq C(\sqrt{\frac{8}{a}})^l \sum_{r=0}^{\infty} (1-\eta)^{r+l} \sum_p (\eta \sqrt{\frac{8}{a}})^p \binom{l+r+p}{p} = \\
&= C(\sqrt{\frac{8}{a}})^l \sum_{r=0}^{\infty} \frac{(1-\eta)^{r+l}}{(1-\eta \sqrt{\frac{8}{a}})^{l+r+1}} = \\
&= (\sqrt{2} - 1)^2 \frac{a^{\frac{5}{2}}}{\eta(\sqrt{a} - \sqrt{8})^2} \left(\frac{\sqrt{8}(1-\eta)}{\sqrt{a} - \eta\sqrt{8}} \right)^l.
\end{aligned}$$

We want that for all $l \Pr(k(t+1) \geq l) \leq a(\frac{1}{a})^l$, so we demand that the factor of the geometrical sequence is smaller than $\frac{1}{a}$ and that the first term is smaller than a . The demands are satisfied with $a = 30$ and $\eta \geq 0.97$. ■

5.2.1 Closing the proof

Theorem 3 *There exists a constant $\eta_1 < 1$ such that for any medium M on n qubits, with decoherence rate $\eta > \eta_1$, which lives for T steps, and any measurement fault F and input string i , the expected cost of $A(M, F)(i)$ is polynomial in n and T .*

proof: We show that for A of mediums with a decoherence rate higher than η_1 defined in lemma 7, and for any qubit q , the sizes of q 's clusters in the algorithm A satisfy for some $a > 8 \Pr(k_q(t) \geq j) \leq \frac{1}{a^{j-1}}$ and $\Pr(K_q^*(t) \geq i) \leq C(\sqrt{\frac{8}{a}})^i$ for $C = (\sqrt{2} - 1)^2 \frac{a^{\frac{3}{2}}}{\sqrt{a} - \sqrt{8}}$. This is done by induction on t . The basis of the induction is trivial, as $K_q(0) = 1, K_q^*(0) = 1$ always. The induction step is done in two stages, by lemmas 6 and 7. With this exponential decaying bound on the distribution over sizes of clusters, we can proceed to show that the expected number of entries of matrices that A writes is polynomial, as in lemma 4, and since each entry can be written using a polynomial number of bits, (if the gates are specified in polynomial number,) this completes the proof, with $\eta_1 = 0.97$. ■

The numerical lower bound of $\eta_1 = 0.97$ is ofcourse not a reasonable physical number, to even start computation with; It is high because it is good for any quantum circuit. Most circuits are simulated efficiently with much lower decoherence rate(see section 5.5).

5.3 The importance of the frequency of many-qubits-gates

A quantum circuit that uses single qubit gates often, is simulated efficiently already for low decoherence rates, since single qubit gates do not participate in the struggle against decoherence. To see that, consider a quantum circuit that applies a many-qubits gate on each qubit once every ν time steps at the most. One can treat these ν time steps as a one time step, in our branching process analysis. During this “extended” time step, the decoherence had it’s chance to operate ν times. The *effective* decoherence rate, i.e. the rate of decoherence with respect to the extended time step, is therefore $1 - (1 - \eta)^\nu \gg \eta$. Therefore already for very small decoherence rates η the effective decoherence will be high enough for the simulation to be efficient.

5.4 The simulation is exponential for low decoherence in random circuits

In this subsection we turn to the other side of the decoherence rate scale, and ask what is the expected cost for low decoherence rates. Off course, there are circuits which can be simulated efficiently even with no decoherence at all. However, for random circuits the expected cost of the simulation is exponential for $\eta < 0.5$. By a random circuit we mean a circuit in which each time step a random matching between qubits is chosen, and this matching corresponds to gates of fan-in two operating on the qubits.

lemma 8 *The simulation of a random circuit subjected to decoherence rate $\eta < 0.5$ is exponential.*

Proof: We prove that when the decoherence rate is $\eta < 0.5$ there is a cluster of size linear in n . During one computation step a cluster of αn qubits, will grow to $(\alpha + \alpha(1 - \alpha))n$ since the probability for qubits in the cluster to be matched outside the cluster is $(1 - \alpha)$, and these matching add at least one qubit to the cluster. The decoherence step multiplies this by η . Hence the original cluster size is multiplied each time step by a factor $(2 - \alpha)\eta$. For $\eta > 0.5$ this is bigger than 1 for small α , and the cluster will grow until it is of size $(2 - \frac{1}{\eta})n$, which is linear in n . The rate of growth of the cluster is exponential, because as long as α is smaller than $2 - \frac{1+\epsilon}{\eta}$, the factor is bigger than $1 + \epsilon$.

5.5 Phase transitions: experimental results

We have performed computer experiments to investigate how the clusters sizes depend on the decoherence rate. To study the random algorithm case, we started with n clusters of size one, and each time step chose a random matching between qubits. A match represented a gate on two qubits, so their clusters were joined. To imitate faults, we randomly and independently separated a qubit from it’s cluster with probability η , each time step. The system exhibits a *phase transition* at decoherence $\eta_0 \approx 0.64$. For $\eta > \eta_0$ the clusters’ sizes stay smaller than $\log(n)$, but for $\eta < \eta_0$ a “giant” cluster of size linear in n appears, which means that the density matrix describing it is exponential.

We also studied experimentally a one dimensional array of qubits, where gates were allowed to operate on nearest neighbors only, and each qubit was matched with it’s left neighbor and right neighbor alternately. We found the transition point at $\eta \approx 0.5$.

6 Conclusions

We feel that this work gives insight on the effects of decoherence on quantum computation, and emphasizes the importance of $O(n)$ parallelism in a quantum computation subjected to decoherence processes.

Our results with respect to sequential quantum computers, and quantum computers with $(\log(n))$ parallelism show that any amount of decoherence destroys all the *quantum* computational power.

Considering parallel computers, an upper bound was given on the decoherence rate which any quantum computer can hope to tolerate. An important role is given to the size of the fan-in used in the computation, and to the frequency of many-qubits-gates. Computers with many single qubit gates seem more sensitive to decoherence, and can be simulated efficiently even with very low decoherence rates. This should be considered when planning implementations of quantum gates by low-fan-in gates[9, 1]. It might also be interesting to quantify, according to these observations, the stability of different algorithms against decoherence.

DiVincenzo[9] raises the question of whether the time evolution of an open quantum system is a powerful computational tool. Our conclusions are that at least within the model we have used, the answer is negative.

An open question is whether a stronger simulation than ours exists, that is efficient for lower decoherence rates or even with no decoherence at all.

Early results[15, 3] indicate the possibility of an efficient simulation of an ideal QC by quantum mediums subjected to low decoherence rates. Though our results might seem as a step towards negating the fascinating possibility of a physical realization of quantum computers, we actually view these results as optimistic, since phase transitions are known to be robust, which implies that maybe quantum computers with low decoherence rates indeed can not be simulated efficiently.

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